

Efficient Multirate Methods from High Order

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Introduction

- Multiphysics applications have specific coupling properties:
 - Coupled in the bulk (magnetohydrodynamics, cosmology)
 - Coupled across interfaces (climate, tokamaks)
- Multiphysics simulation challenges include:
 - Multirate processes, but too close to analytically reformulate.
 - Optimal solvers may exist for some pieces, but not for the whole.
 - Mixing of stiff/nonstiff processes, challenging standard solvers.
- Many legacy codes utilize lowest-order time step splittings, may suffer from:
 - Low accuracy typically $O(h)$ -accurate; symmetrization/extrapolation may improve this but at significant cost [Ropp, Shadid & Ober 2005].
 - Poor/unknown stability even when each part utilizes a stable step size, the combined problem may admit unstable modes [Estep et al., 2007].

Need for Flexible High-Order Multirate Integrators

Multirate methods evolve distinct problem components with their own rate-specific time steps.

- Historical approaches:
 - Simple (h)-accurate subcycling approaches
 - Interpolation to handle fast/slow coupling (typically (h^2), sometimes (h^3)) [Kværnø & Rentrop, 1999; . . .].
 - Extrapolation methods to bootstrap accuracy for low order methods [Engstler & Lubich, 1997; Constantinescu & Sandu, 2013; . . .].
- Next-generation methods will require a variety of criteria:
 - High-order accuracy stability, both within and between components
 - Flexible rate structure within integration, or even to dynamically identify fast vs slow partitioning of components
 - Robust temporal error estimation adaptivity of step size(s)
 - Enable problem-specific options, e.g. SSP or symplectic for specific components

GARK framework for 2-rate problem

- Consider:

$\mathbf{A}^{\{f,f\}}$	$\mathbf{A}^{\{f,s\}}$
$\mathbf{A}^{\{s,f\}}$	$\mathbf{A}^{\{s,s\}}$
$\mathbf{b}^{\{f\}\tau}$	$\mathbf{b}^{\{s\}\tau}$
- General: $\frac{dy}{dt} = F(t, y), \quad y(t_0) = y_0$
 - $F(t, y)$ with fast portion and a slow portion
 - slow time-step size h , fast time-step size h/m
 - time-scale separation m

- Additive:

$$y'(t) = F(t, y), \quad F(t, y) = f_f(t, y) + f_s(t, y), \quad t \geq t_0, \quad y(t_0) = y_0$$

- Stage and Solution Updates:

$$\mathbf{k}_j^{\{f\}} = \mathbf{y}_n + h \sum_{l=1}^{\mathbf{s}\{f\}} a_{j,l}^{\{f,f\}} f^{\{f\}}(t_n + c_l^{\{f\}}h, \mathbf{k}_l^{\{f\}}) + h \sum_{l=1}^{\mathbf{s}\{s\}} a_{j,l}^{\{f,s\}} f^{\{s\}}(t_n + c_l^{\{s\}}h, \mathbf{k}_l^{\{s\}})$$

$$\mathbf{k}_i^{\{s\}} = \mathbf{y}_n + h \sum_{l=1}^{\mathbf{s}\{f\}} a_{i,l}^{\{s,f\}} f^{\{f\}}(t_n + c_l^{\{f\}}h, \mathbf{k}_l^{\{f\}}) + h \sum_{l=1}^{\mathbf{s}\{s\}} a_{i,l}^{\{s,s\}} f^{\{s\}}(t_n + c_l^{\{s\}}h, \mathbf{k}_l^{\{s\}})$$

$$\mathbf{y}_{n+1} = \mathbf{y}_n + h \sum_{l=1}^{\mathbf{s}\{f\}} b_l^{\{f\}} f^{\{f\}}(t_n + c_l^{\{f\}}h, \mathbf{k}_l^{\{f\}}) + h \sum_{l=1}^{\mathbf{s}\{s\}} b_l^{\{s\}} f^{\{s\}}(t_n + c_l^{\{s\}}h, \mathbf{k}_l^{\{s\}})$$

- Row-sum conditions give stage times: $c_j^{\{f\}} = \sum_{l=1}^{\mathbf{s}\{f\}} a_{jl}^{\{f,f\}} = \sum_{l=1}^{\mathbf{s}\{s\}} a_{jl}^{\{f,s\}}$ $c_j^{\{s\}} = \sum_{l=1}^{\mathbf{s}\{s\}} a_{jl}^{\{s,s\}} = \sum_{l=1}^{\mathbf{s}\{f\}} a_{jl}^{\{s,f\}}$

GARK h^4 Order Conditions

For $\sigma, \nu, \mu \in \{f, s\}$, and assuming $\mathbf{c}^{\{\sigma\}} = \mathbf{A}^{\{\sigma, f\}} \mathbf{1}^{\{f\}} = \mathbf{A}^{\{\sigma, s\}} \mathbf{1}^{\{s\}}$:

$$\mathbf{b}^{\{\sigma\} \top} \mathbf{1}^{\{\sigma\}} = 1, \quad \mathbf{b}^{\{\sigma\} \top} \mathbf{c}^{\{\sigma\}} = \frac{1}{2} \quad [h, h^2]$$

$$\mathbf{b}^{\{\sigma\} \top} \left(\mathbf{c}^{\{\sigma\}} \right)^2 = \frac{1}{3}, \quad \mathbf{b}^{\{\sigma\} \top} \mathbf{A}^{\{\sigma, \nu\}} \mathbf{c}^{\{\nu\}} = \frac{1}{6} \quad [h^3]$$

$$\mathbf{b}^{\{\sigma\} \top} \left(\mathbf{c}^{\{\sigma\}} \right)^3 = \frac{1}{4}, \quad \left(\mathbf{b}^{\{\sigma\}} \times \mathbf{c}^{\{\sigma\}} \right)^\top \mathbf{A}^{\{\sigma, \nu\}} \mathbf{c}^{\{\nu\}} = \frac{1}{8} \quad [h^4]$$

$$\mathbf{b}^{\{\sigma\} \top} \mathbf{A}^{\{\sigma, \nu\}} \left(\mathbf{c}^{\{\nu\}} \right)^2 = \frac{1}{12}, \quad \mathbf{b}^{\{\sigma\} \top} \mathbf{A}^{\{\sigma, \mu\}} \mathbf{A}^{\{\mu, \nu\}} \mathbf{c}^{\{\nu\}} = \frac{1}{24}.$$

Here, exponentiation and \times denote component-wise operators.

We'll refer to these as “fast” conditions when $\sigma = f$ (and “slow” when $\sigma = s$).

As expected, the number of conditions increases dramatically with order: 2 for h , 4 for h^2 , 10 for h^3 , and 28 for h^4 (note: h^5 has 86).

MIS methods

- GARK: flexible theory for solving order conditions
- Construct from base inner and outer methods $T_O = \{A^O, b^O, c^O\}$, where $c_i^O \leq c_{i+1}^O$, $i = 1, \dots, s^O - 1$ and $T_I = \{A^I, b^I, c^I\}$, where $c_i^I \leq c_{i+1}^I$, $i = 1, \dots, s^I$.
- MIS method formulation solves sub-problem [1]
- RFSMR concept focuses on defining the residual for splitting [4]
 - $r_i = \sum_{j=1}^{i-1} (a_{ij}^O - a_{i-1,j}^O) f^{\{s\}}(\mathbf{k}_j^{\{s\}})$
 - $\frac{\partial \mathbf{k}^{\{f,i\}}}{\partial \tau} = \frac{1}{c_i^O - c_{i-1}^O} r_i + f^{\{f\}}(\mathbf{k}^{\{f,i\}})$, $\tau \in [\tau_{i,1}, \tau_{i+1,1}]$, $i = 2, \dots, s^O + 1$
 - Final step solution accumulated similarly to stage solutions
- If both T_O and T_I are at least h^3 , and T^O satisfies

$$\sum_{i=2}^{s^O} (c_i^O - c_{i-1}^O) (e_i + e_{i-1})^\top A^O c^O + (1 - c_{s^O}^O) \left(\frac{1}{2} + e_{s^O}^\top A^O c^O \right) = \frac{1}{3}, \quad (1)$$

then the MIS method is h^3 .

Relaxed Multirate Infinitesimal Step (RMIS) Methods compared

- New method: RMIS

- Uses same sub-problems as MIS

- $r_i = \sum_{j=1}^{i-1} (a_{ij}^O - a_{i-1,j}^O) f^{\{s\}}(\mathbf{k}_j^{\{s\}})$

- $\frac{\partial \mathbf{k}^{\{f,i\}}}{\partial \tau} = \frac{1}{c_i^O - c_{i-1}^O} r_i + f^{\{f\}}(\mathbf{k}^{\{f,i\}}), \quad \tau \in [\tau_{i,1}, \tau_{i+1,1}], i = 2, \dots, s^O + 1$

- Preserves linear invariants

- Final step solution accumulated by using only fast stage values at the stage times the slow function is evaluate

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 - Preserves linear invariants
 - Final step solution accumulated by using only fast stage values at the stage times the slow function is evaluate
- Comparatively, choose:

$$\mathbf{b}_{\text{MIS}}^{\{f\}\top} = [c_2^O b^{I\top} \quad (c_3^O - c_2^O) b^{I\top} \quad \dots \quad (1 - c_{s^O}^O) b^{I\top}]$$

$$\mathbf{b}_{\text{RMIS}}^{\{f\}\top} = [b_1^O \mathbf{e}_1^\top \quad b_2^O \mathbf{e}_1^\top \quad \dots \quad b_{s^O}^O \mathbf{e}_1^\top] \in \mathbb{R}^{s^O s^I} = \mathbb{R}^{s^f}.$$

$$\text{where } \mathbf{e}_1^\top = [1 \quad 0 \quad \dots \quad 0] \in \mathbb{Z}^{s^I}$$

Simplification of RMIS Order Conditions

Lemma (Sexton & Reynolds, 2018)

Choosing $\mathbf{b}^{\{f\}} = \mathbf{b}_{RMIS}^{\{f\}}$, and assuming T_I has explicit first stage, then:

$$\begin{aligned}\mathbf{b}^{\{f\}\top} \left(\mathbf{c}^{\{f\}} \right)^q &= \mathbf{b}^{\{s\}\top} \left(\mathbf{c}^{\{s\}} \right)^q, \quad q = 0, 1, \dots, \\ \mathbf{b}^{\{f\}\top} \mathbf{A}^{\{f,f\}} &= \mathbf{b}^{\{s\}\top} \mathbf{A}^{\{s,f\}}, \\ \mathbf{b}^{\{f\}\top} \mathbf{A}^{\{f,s\}} &= \mathbf{b}^{\{s\}\top} \mathbf{A}^{\{s,s\}}, \\ \left(\mathbf{b}^{\{f\}} \times \mathbf{c}^{\{f\}} \right)^\top \mathbf{A}^{\{f,f\}} &= \left(\mathbf{b}^{\{s\}} \times \mathbf{c}^{\{s\}} \right)^\top \mathbf{A}^{\{s,f\}}, \\ \left(\mathbf{b}^{\{f\}} \times \mathbf{c}^{\{f\}} \right)^\top \mathbf{A}^{\{f,s\}} &= \left(\mathbf{b}^{\{s\}} \times \mathbf{c}^{\{s\}} \right)^\top \mathbf{A}^{\{s,s\}}.\end{aligned}$$

Hence, all of the fast fourth order conditions are equivalent to their slow counterparts, reducing the 28 total conditions to just 14.

We anticipate a similar result for the fifth order conditions ($86 \rightarrow 43$), but have yet to perform the analysis.

RMIS Method Order

Theorem (Sexton & Reynolds, 2018)

Assume that T_I is at least third order. Assume that T_O is explicit, at least fourth order, and satisfies

$$v^{O\top} A^O c^O = \frac{1}{12}, \quad (2)$$

where

$$v_i^O = \begin{cases} 0, & i = 1, \\ b_i^O (c_i^O - c_{i-1}^O) + (c_{i+1}^O - c_{i-1}^O) \sum_{j=i+1}^{s^O} b_j^O, & 1 < i < s^O, \\ b_{s^O}^O (c_{s^O}^O - c_{s^O-1}^O), & i = s^O, \end{cases}$$

then the coefficients $\mathbf{A}^{\{f,f\}}$, $\mathbf{A}^{\{f,s\}}$, $\mathbf{A}^{\{s,f\}}$, $\mathbf{A}^{\{s,s\}}$ and $\mathbf{b}^{\{s\}}$ satisfy all of the “slow” fourth-order conditions.

Condition (2) is analogous to (1), that guarantees the MIS method is h^3 .

RMIS & MIS Order Summary

Combining these two results with the existing MIS method theory, we have:

MIS: if (a) T_I is h^3 , (b) T_O is explicit and h^3 , and (c) T_O satisfies (1), then the MIS method is h^3 .

RMIS: if (a) T_I is h^3 and has explicit first stage, (b) T_O is explicit and h^4 , and (c) T_O satisfies (2), then the RMIS method is h^4 .

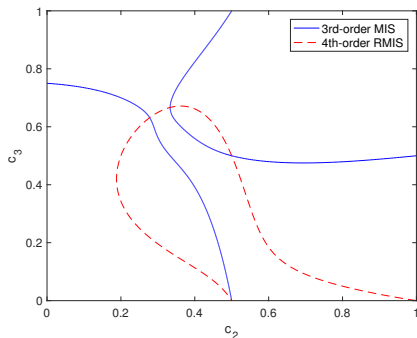
Finally, since MIS and RMIS only differ in their selection of $\mathbf{b}^{\{f\}}$, then if all of the above assumptions are satisfied, we may use MIS as an h^3 *embedding* for the h^4 RMIS method.

Choosing Base Methods

- Represent 4-stage 4th order RK method in terms of stage times c_2 and c_3
- Solve RFSMR and RMIS order condition on the outer/slow base method

$$3/8 - \text{Rule} : \begin{array}{c|cccc} 0 & 0 & 0 & 0 & 0 \\ 1 & \frac{1}{3} & 0 & 0 & 0 \\ \frac{2}{3} & -\frac{1}{3} & 1 & 0 & 0 \\ \frac{3}{3} & 1 & -1 & 1 & 0 \\ \hline & \frac{1}{8} & \frac{3}{8} & \frac{3}{8} & \frac{1}{8} \end{array}$$

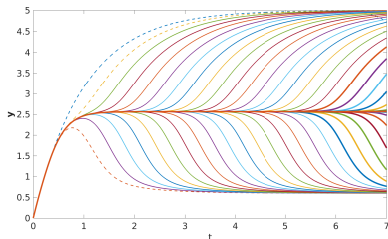
$$\text{KW3} : \begin{array}{c|ccc} 0 & 0 & 0 & 0 \\ \frac{1}{3} & \frac{1}{3} & 0 & 0 \\ \frac{3}{3} & -\frac{3}{16} & \frac{15}{16} & 0 \\ \frac{4}{4} & \frac{1}{6} & \frac{3}{10} & \frac{8}{15} \end{array}$$



Numerical order and efficiency results

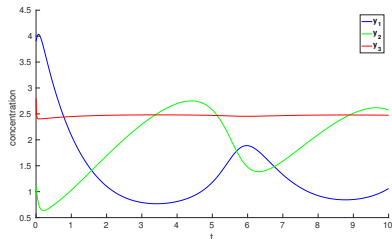
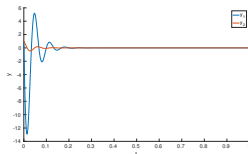
- Test Problems

- Inverter-chain:
weakly coupled,
literature [3]
- Kuhn stability:
strongly coupled,
linear [2]
- Brusselator:
chemical reaction
network,
nonlinear [?]

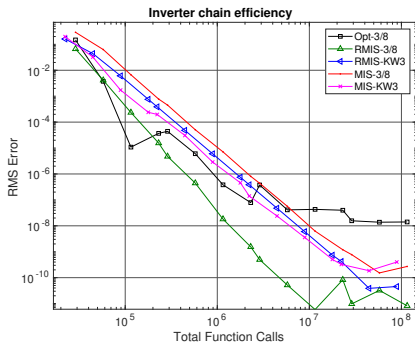
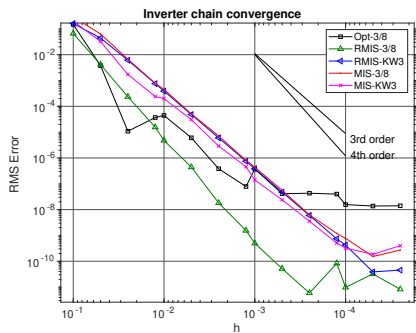


- Methods

- RMIS w/ 4-stage Base (4th)
- RMIS w/ 3-stage Knoth-Wolke (3rd)
- RFSMR w/ 4-stage Base (3rd)
- RFSMR w/ 3-stage Knoth-Wolke (3rd)

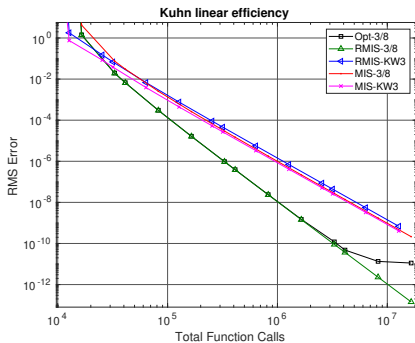
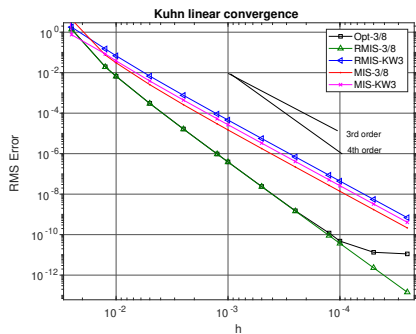


Inverter-chain test results



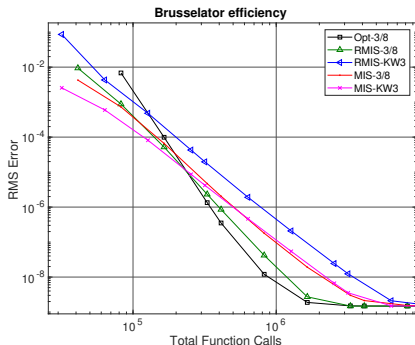
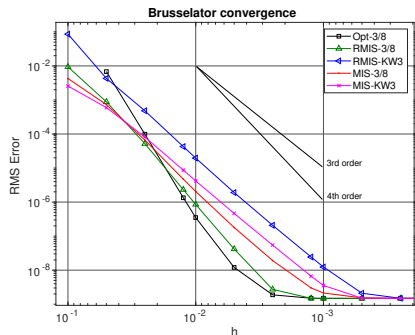
- Fixed step h
- Efficiency horizontal shift depends on number of stages in base method
- RMS error $\sqrt{\sum_{i=1}^n \frac{(\hat{y}_i - y_i)^2}{n}}$ where \hat{y} from high order implicit solve with tiny h

Kuhn stability test results

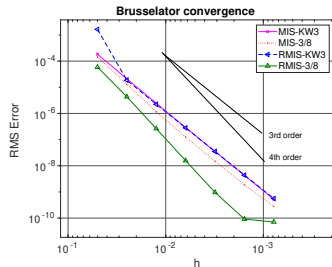
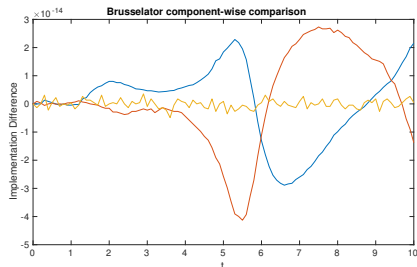


- Fixed step h
- Numerical order and efficiency results are cleaner for this 2×2 linear problem
- RMS error $\sqrt{\sum_{i=1}^n \frac{(\mathbf{y}_{\text{true}_i} - \mathbf{y}_i)^2}{n}}$ where \mathbf{y}_{true} is exact solution

Brusselator test results



- With fixed step h , our new methods are more efficient for stronger error requirements
- RMS error $\sqrt{\sum_{i=1}^n \frac{(\hat{y}_i - y_i)^2}{n}}$ where \hat{y} from high order implicit solve with tiny h



- An updated implementation can be found at <https://drreynolds@bitbucket.org/drreynolds/rmis.git>
- Testing the RMIS-3/8 method with this new implementation on the Brusselator problem shows close to round-off absolute error differences
- The same convergence properties are observed

Conclusions

- The Generalized-structure Additively-partitioned Runge Kutta can be used in creating new methods based on Multirate Infinitesimal Step Methods with desirable properties
- Using one of our coupling approaches with a base method that also satisfies the slow coupling conditions is a fourth order overall method.
- These multiple coupling approaches allows for approximations of local error by using them together.
- Future areas of interest include:
 - Time-step adaptivity for the slow-time scale based on embeddings
 - Time-step adaptivity for the time-scale ratio based on embeddings
 - Investigate extensions to allow implicitness at the slow time scale
 - Extensions to fifth order (or higher)

Acknowledgements & Questions

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GARK Fifth Order Conditions

The h^5 conditions are, for $\sigma, \nu, \mu, \lambda \in f, s$:

$$\begin{aligned} \mathbf{b}^{\{\sigma\}\top} \left(\mathbf{c}^{\{\sigma\}} \right)^4 &= \frac{1}{5} && [2 \text{ conditions}], \\ \left(\mathbf{b}^{\{\sigma\}} \times \left(\mathbf{c}^{\{\sigma\}} \right)^2 \right)^\top \mathbf{A}^{\{\sigma, \nu\}} \mathbf{c}^{\{\nu\}} &= \frac{1}{10} && [4 \text{ conditions}], \\ \left(\mathbf{b}^{\{\sigma\}} \times \mathbf{c}^{\{\sigma\}} \right)^\top \mathbf{A}^{\{\sigma, \nu\}} \left(\mathbf{c}^{\{\nu\}} \right)^2 &= \frac{1}{15} && [4 \text{ conditions}], \\ \left(\mathbf{b}^{\{\sigma\}} \times \mathbf{c}^{\{\sigma\}} \right)^\top \mathbf{A}^{\{\sigma, \nu\}} \mathbf{A}^{\{\nu, \mu\}} \mathbf{c}^{\{\mu\}} &= \frac{1}{30} && [8 \text{ conditions}], \\ \mathbf{b}^{\{\sigma\}\top} \left(\mathbf{A}^{\{\sigma, \nu\}} \mathbf{c}^{\{\nu\}} \right)^2 &= \frac{1}{20} && [4 \text{ conditions}], \\ \mathbf{b}^{\{\sigma\}\top} \mathbf{A}^{\{\sigma, \nu\}} \left(\mathbf{c}^{\{\nu\}} \right)^3 &= \frac{1}{20} && [4 \text{ conditions}], \\ \mathbf{b}^{\{\sigma\}\top} \mathbf{A}^{\{\sigma, \nu\}} \left(\mathbf{c}^{\{\nu\}} \times \left(\mathbf{A}^{\{\nu, \mu\}} \mathbf{c}^{\{\mu\}} \right) \right) &= \frac{1}{40} && [8 \text{ conditions}], \\ \mathbf{b}^{\{\sigma\}\top} \mathbf{A}^{\{\sigma, \nu\}} \mathbf{A}^{\{\nu, \mu\}} \left(\mathbf{c}^{\{\mu\}} \right)^2 &= \frac{1}{60} && [8 \text{ conditions}], \\ \mathbf{b}^{\{\sigma\}\top} \mathbf{A}^{\{\sigma, \nu\}} \mathbf{A}^{\{\nu, \mu\}} \mathbf{A}^{\{\mu, \lambda\}} \mathbf{c}^{\{\lambda\}} &= \frac{1}{120} && [16 \text{ conditions}]. \end{aligned}$$

Subcycling as a One-Step Method

Consider taking 3 substeps of size $\frac{h}{3}$ with the midpoint method,

$$\begin{array}{c|cc} 0 & 0 & 0 \\ \frac{1}{2} & \frac{1}{2} & 0 \\ \hline & 0 & 1 \end{array} = \frac{c}{b^\top} A:$$

Basic steps

$$z_1 = y_n$$

$$z_2 = y_n + \frac{h}{6} f(z_1)$$

$$y^* = y_n + \frac{h}{3} f(z_2)$$

$$z_3 = y^*$$

$$z_4 = y^* + \frac{h}{6} f(z_3)$$

$$y^{**} = y^* + \frac{h}{3} f(z_4)$$

$$z_5 = y^{**}$$

$$z_6 = y^{**} + \frac{h}{6} f(z_5)$$

Single step

$$z_1 = y_n$$

$$z_2 = y_n + \frac{h}{6} f(z_1)$$

$$z_3 = y_n + \frac{h}{3} f(z_2)$$

$$z_4 = y_n + \frac{h}{3} f(z_2) + \frac{h}{6} f(z_3)$$

$$z_5 = y_n + \frac{h}{3} f(z_2) + \frac{h}{3} f(z_4)$$

$$z_6 = y_n + \frac{h}{3} f(z_2) + \frac{h}{3} f(z_4)$$

$$+ \frac{h}{6} f(z_5)$$

$$y_{n+1} = y_n + \frac{h}{3} f(z_2) + \frac{h}{3} f(z_4)$$

$$+ \frac{h}{6} f(z_5)$$

Butcher tableau

0	0	0	0	0	0	0
$\frac{1}{6}$	$\frac{1}{6}$	0	0	0	0	0
$\frac{2}{6}$	0	$\frac{1}{3}$	0	0	0	0
$\frac{3}{6}$	0	$\frac{1}{3}$	$\frac{1}{6}$	0	0	0
$\frac{4}{6}$	0	$\frac{1}{3}$	0	$\frac{1}{3}$	0	0
$\frac{5}{6}$	0	$\frac{1}{3}$	0	$\frac{1}{3}$	$\frac{1}{6}$	0
<hr/>						
	0	$\frac{1}{3}$	0	$\frac{1}{3}$	0	$\frac{1}{3}$

\Downarrow

$\frac{1}{3}c$	$\frac{1}{3}A$	0	0
$\frac{1}{3}1 + \frac{1}{3}c$	$\frac{1}{3}1b^\top$	$\frac{1}{3}A$	0
$\frac{2}{3}1 + \frac{1}{3}c$	$\frac{1}{3}1b^\top$	$\frac{1}{3}1b^\top$	$\frac{1}{3}A$